

# Homework Assignment 2

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**Note:** Homework was discussed with Varun Miranda and Bhushan Malgaonkar

**Ans 1.**

$$f(x/\alpha) = \alpha^x(1 - \alpha)^{1-x}$$

Given  $\alpha$  is selected from a uniform distribution(0,1)

$$\text{Therefore } f(\alpha) = \frac{1}{1-0} = 1$$

$$\operatorname{argmax}_{\alpha}(f(\alpha)f(x/\alpha)) = 1. \prod_{i=1}^n \alpha^x(1 - \alpha)^{1-x}$$

Taking log

$$= \sum_{i=1}^n (x \log(\alpha) + (1 - x) \log(1 - \alpha))$$

Taking derivative w.r.t  $\alpha$

$$0 = \frac{1}{\alpha} \sum_{i=1}^n x - \frac{1}{1 - \alpha} \sum_{i=1}^n (1 - x)$$

$$\frac{1}{\alpha} \sum_{i=1}^n x = \frac{1}{1 - \alpha} \sum_{i=1}^n (1 - x)$$

$$\sum_{i=1}^n x - \alpha \sum_{i=1}^n x = N\alpha - \alpha \sum_{i=1}^n x$$

$$\hat{\alpha}_{MAP} = \frac{1}{N} \sum_{i=1}^n x = \bar{x}$$

**Ans 2.**

$$\operatorname{argmax}_{\theta} \prod_{i=0}^n \theta e^{-\theta x}$$

Taking log

$$\sum_{i=0}^n \log \theta - \theta x$$

Taking derivative w.r.t.  $\theta$

$$\sum_{i=0}^n \left( \frac{1}{\theta} - x_i \right) = 0$$

$$\frac{n}{\theta} = \sum_{i=0}^n x_i$$

$$\hat{\theta}_{MLE} = \frac{1}{\bar{x}}$$

**Ans 3. a).**

For  $H_0$ ,

$$f_X(x) = f_N\left(\frac{x+s}{a}\right)$$

For  $H_1$ ,

$$f_X(x) = f_N\left(\frac{x-s}{b}\right)$$

$$f_N\left(\frac{x+s}{a}\right) \underset{H_1}{\overset{H_0}{\geq}} f_N\left(\frac{x-s}{b}\right)$$

Since  $a = b$

$$f_N\left(\frac{x+s}{a}\right) \underset{H_1}{\overset{H_0}{\geq}} f_N\left(\frac{x-s}{a}\right)$$

$$\frac{1}{\pi(1 + (\frac{x+s}{a})^2)} \underset{H_1}{\overset{H_0}{\geq}} \frac{1}{\pi(1 + (\frac{x-s}{a})^2)}$$

$$1 + \left(\frac{x-s}{a}\right)^2 \underset{H_1}{\overset{H_0}{\geq}} 1 + \left(\frac{x+s}{a}\right)^2$$

$$x^2 - 2xs + s^2 \underset{H_1}{\overset{H_0}{\geq}} x^2 + 2xs + s^2$$

$$0 \underset{H_1}{\overset{H_0}{\geq}} 4xs$$

Since  $s > 0$ ,  $s$  is positive

$$0 \underset{H_1}{\overset{H_0}{\geq}} x$$

□

b).

$$r(\hat{y}) = \int_{H0} f(x|y = H1).P(y = H1)dx + \int_{H1} f(x|y = H0).P(y = H0)dx$$

Since priors are equal  $P(y = H1) = P(y = H0) = \frac{1}{2}$

$$\begin{aligned} r(\hat{y}) &= \frac{1}{2} \int_{H0} f(x|y = H1)dx + \int_{H1} f(x|y = H0)dx \\ &= \frac{1}{2} \int_{-\infty}^0 f_N\left(\frac{x+s}{a}\right) + \frac{1}{2} \int_0^{\infty} f_N\left(\frac{x-s}{b}\right) \\ &= \frac{1}{2} \int_{-\infty}^0 \frac{1}{\pi(1 + (\frac{x+s}{a})^2)} + \frac{1}{2} \int_0^{\infty} \frac{1}{\pi(1 + (\frac{x-s}{b})^2)} \\ &= \frac{1}{2\pi} \int_{-\infty}^0 \frac{1}{1 + (\frac{x+s}{a})^2} + \frac{1}{2\pi} \int_0^{\infty} \frac{1}{1 + (\frac{x-s}{b})^2} \end{aligned}$$

put  $t = \frac{x+s}{a}$  and  $u = \frac{x-s}{b}$

When  $x = -\infty$ ,  $t = -\infty$ . When  $x = 0$ ,  $t = \frac{s}{a}$

When  $x = 0$ ,  $u = \frac{-s}{b}$ . When  $x = \infty$ ,  $u = \infty$

$$\begin{aligned} r(\hat{y}) &= \frac{1}{2\pi} \int_{-\infty}^{s/a} \frac{1}{1+t^2} dt + \frac{1}{2\pi} \int_{-s/b}^{\infty} \frac{1}{1+u^2} du \\ &= \frac{1}{2\pi} (\tan^{-1}(\frac{s}{a}) + \frac{\pi}{2} + \frac{\pi}{2} - \tan^{-1}(\frac{-s}{b})) \\ &= \frac{1}{2} + \frac{1}{2\pi} (\tan^{-1}(\frac{s}{a}) - \tan^{-1}(\frac{-s}{b})) \end{aligned}$$

□

Ans 4. a).

$$\begin{aligned}
 f_{Y|X}(y|x) &= \frac{f_{X|Y}(x|y) \cdot f_Y(y)}{f_X(x)} \\
 f_Y(y) &= \alpha e^{-y\alpha} \\
 f_X(x) &= \int_0^\infty f_{X|Y}(x|y) \cdot f_Y(y) dy \\
 &= \int_0^\infty y e^{-yx} \cdot \alpha e^{-y\alpha} dy \\
 &= \alpha \int_0^\infty y e^{-y(x+\alpha)} dy
 \end{aligned}$$

Using Integration by parts and LIATE Rule

$$\begin{aligned}
 &= \alpha \left( y \int e^{-y(x+\alpha)} dy - \int \left( \frac{d(y)}{dy} \int e^{-y(x+\alpha)} dy \right) dy \right) \\
 &= \alpha \left[ \frac{-y e^{-y(x+\alpha)}}{x+\alpha} - \frac{e^{-y(x+\alpha)}}{(x+\alpha)^2} \right]_0^\infty \\
 &= \alpha \left[ \frac{-e^{-y(x+\alpha)}}{x+\alpha} \left( y + \frac{1}{x+\alpha} \right) \right]_0^\infty \\
 &= \alpha \left[ \frac{e^{-y(x+\alpha)}}{x+\alpha} \left( y + \frac{1}{x+\alpha} \right) \right]_\infty^0 \\
 &= \alpha \left[ \frac{1}{x+\alpha} \left( 0 + \frac{1}{x+\alpha} \right) - 0 \right] \\
 &= \frac{\alpha}{(x+\alpha)^2} \\
 f_{Y|X}(y|x) &= \frac{y e^{-yx} \alpha e^{-y\alpha}}{\frac{\alpha}{(x+\alpha)^2}} \\
 &= (x+\alpha)^2 \cdot y \cdot e^{-y(x+\alpha)}
 \end{aligned}$$

□

b).

□

c).

□

**Ans 5. a).**

$$P(S = 1|V = 1) = P(S = 1) \text{ (Since S is independent of V)}$$

$$\begin{aligned} P(S = 1) &= P(S = 1|G = 0).P(G = 0) + P(S = 1|G = 1).P(G = 1) \\ &= (1 - \gamma)\alpha + (1 - \beta)(1 - \alpha) \end{aligned}$$

□

$$b). P(S=1|V=0) = P(S=1) \text{ (Since S is independent of V)}$$

$$= P(S=1|V=1)$$

Explanation: S is independent of V. Hence value of V doesn't affect S

□

Ans 6. a).

$$\begin{aligned}P(X_2 = \text{salmon}) &= P(X_1 = \text{Winter}) \cdot P(X_2 = \text{salmon} | X_1 = \text{Winter}) + \\&\quad P(X_1 = \text{Autumn}) \cdot P(X_2 = \text{salmon} | X_1 = \text{Autumn}) \\&= 0.5 * 0.9 + 0.5 * 0.8 \\&= 0.85 \\P(X_2 = \text{seabass}) &= P(X_1 = \text{Winter}) \cdot P(X_2 = \text{seabass} | X_1 = \text{Winter}) + \\&\quad P(X_1 = \text{Autumn}) \cdot P(X_2 = \text{seabass} | X_1 = \text{Autumn}) \\&= 0.5 * 0.1 + 0.5 * 0.2 \\&= 0.15\end{aligned}$$

□

b).

$$\begin{aligned}P(\text{Winter}) &= P(X_1 = \text{Winter}) * P(X_2 = \text{salmon} | X_1 = \text{Winter}) * P(X_3 = \text{Dark} | X_2 = \text{salmon}) \\&\quad * P(X_4 = \text{Wide} | X_2 = \text{salmon}) + P(X_1 = \text{Winter}) * P(X_2 = \text{seabass} | X_1 = \text{Winter}) \\&\quad * P(X_3 = \text{Dark} | X_2 = \text{seabass}) * P(X_4 = \text{Wide} | X_2 = \text{seabass}) \\&= 0.25(0.9 * 0.34 * 0.4 + 0.1 * 0.1 * 0.95) \\&= 0.0323 \\P(\text{Spring}) &= P(X_1 = \text{Spring}) * P(X_2 = \text{salmon} | X_1 = \text{Spring}) * P(X_3 = \text{Dark} | X_2 = \text{salmon}) \\&\quad * P(X_4 = \text{Wide} | X_2 = \text{salmon}) + P(X_1 = \text{Spring}) * P(X_2 = \text{seabass} | X_1 = \text{Spring}) \\&\quad * P(X_3 = \text{Dark} | X_2 = \text{seabass}) * P(X_4 = \text{Wide} | X_2 = \text{seabass}) \\&= 0.25(0.3 * 0.34 * 0.4 + 0.7 * 0.1 * 0.95) \\&= 0.0268 \\P(\text{Summer}) &= P(X_1 = \text{Summer}) * P(X_2 = \text{salmon} | X_1 = \text{Summer}) * P(X_3 = \text{Dark} | X_2 = \text{salmon}) \\&\quad * P(X_4 = \text{Wide} | X_2 = \text{salmon}) + P(X_1 = \text{Summer}) * P(X_2 = \text{seabass} | X_1 = \text{Summer}) \\&\quad * P(X_3 = \text{Dark} | X_2 = \text{seabass}) * P(X_4 = \text{Wide} | X_2 = \text{seabass}) \\&= 0.25(0.4 * 0.34 * 0.4 + 0.6 * 0.1 * 0.95) \\&= 0.02785 \\P(\text{Autumn}) &= P(X_1 = \text{Autumn}) * P(X_2 = \text{salmon} | X_1 = \text{Autumn}) * P(X_3 = \text{Dark} | X_2 = \text{salmon}) \\&\quad * P(X_4 = \text{Wide} | X_2 = \text{salmon}) + P(X_1 = \text{Autumn}) * P(X_2 = \text{seabass} | X_1 = \text{Autumn}) \\&\quad * P(X_3 = \text{Dark} | X_2 = \text{seabass}) * P(X_4 = \text{Wide} | X_2 = \text{seabass}) \\&= 0.25(0.8 * 0.34 * 0.4 + 0.2 * 0.1 * 0.95) \\&= 0.03195\end{aligned}$$

It is most likely to be winter

□

**Ans 7.**

**Ans 8.** *b*).

